



Secure Routine: A Routine-Based Algorithm for Drivers Identification

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Who am I?

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Routine-
Based
Algorithm
for Drivers
Identification



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Goal

Secure
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Identification

Identify driver and distinguish the vehicle's **owner** from other drivers exploiting **sensors' vehicle data**



Why?

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- *Theft detection.* In 2019, around 56k vehicles were stolen in UK
- *Insurance companies*



How to identify drivers?

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Identification is possible with classification of driver's behavioural.

Routine based classification

Classification based on how frequently actions are repeated in time.

Two persons may complete a task with similar action but with few important variations



Secure Routine

Panoramic

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- Secure Routine (SR) learns what drivers do and how much frequently
- SR algorithm reads sensors' car data via OBD-II interface
- SR algorithm elaborates recorded data and extracts from timestamp second, minute, hours, day of week, day, month, year
- SR make use of Machine Learning (ML) algorithms to discover driver's routine and to identify the driver



Secure Routine

What differs from other identification solutions?

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- Most identification algorithms do not look for frequency
- Previous solutions use same feature set for all drivers
- Many algorithms ignore timestamp component

Instead:

- SR selects best feature set for each driver
- SR decomposes timestamp in fine grained units



Secure Routine

Example

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Alice and Bob are used to going on the Sixth Avenue. Alice usually take the road at 12PM and Bob at 7PM. If the vehicle is at 7PM on the Sixth Avenue, who is the driver?

- Without timestamp, the resulting model cannot answer unambiguously
- With timestamp, the resulting model says "The driver is Bob"



Secure Routine Algorithm

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- *Model Generation Dataset* creates instances composed by (day, month, year, hour, minute, second and day of the week) plus sensors' car data
- *Feature Selection Paradigm* ranks the features applying the Gain Ratio and selects the best feature
- *Model Generation Algorithm* uses past instances to generate a vehicle's driver model
- *SR Identification strategy* evaluates if each instance belongs to vehicle's owner or to another driver



Secure Routine Evaluation

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- ¹ referred in the following as M
- ² referred in the following as K
- ³ referred in the following as G

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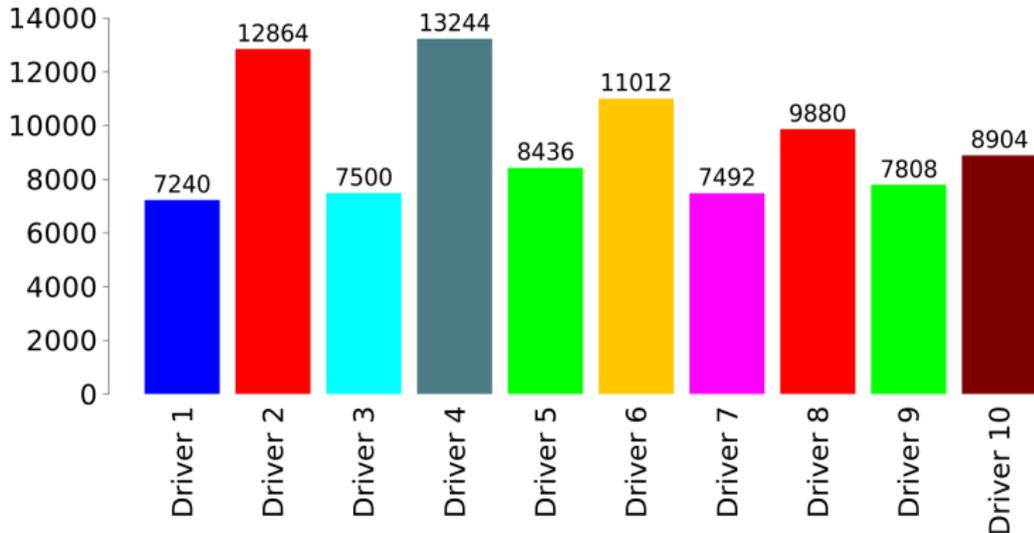


Secure Routine Evaluation

Datasets

Secure Routine: A Routine-Based Algorithm for Drivers Identification

Dataset Θ :



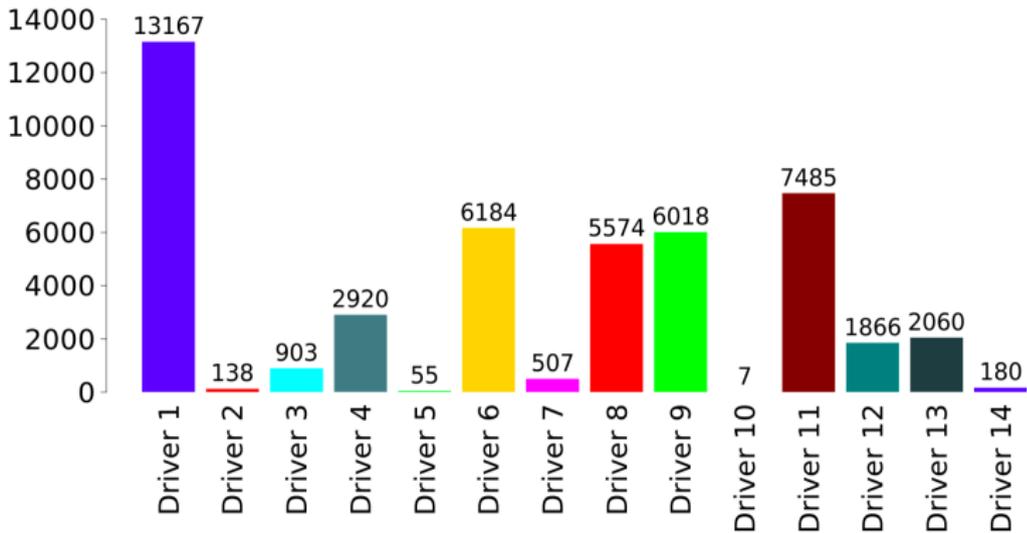


Secure Routine Evaluation

Datasets

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Dataset Ψ :





Experiments

Comparison of Secure Routine with M, K and G on Θ

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Secure Routine		<i>M</i>		<i>K</i>		<i>G</i>	
Precision	Recall	Precision	Recall	Precision	Recall	Precision	Recall
99,6%	99,6%	99,2%	99,2%	N.A.	N.A.	98,8%	98,1%
Accuracy		Accuracy		Accuracy		Accuracy	
99,6%		99,2%		99,6%		N.A.	



Experiments

Comparison of Secure Routine with M on Θ

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Secure Routine		M	
Avg. Precision	Avg. Recall	Avg. Precision	Avg. Recall
99,8%	98,5%	99,3%	99,3%



Experiments

Comparison of Secure Routine with M for multi-driver identification on Ψ

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Secure Routine		M	
Precision	Recall	Precision	Recall
99,4%	99,4%	90,4%	89,8%



Experiments

Comparison of Secure Routine with M for owner identification on Ψ

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Secure Routine		M	
Avg. Precision	Avg. Recall	Avg. Precision	Avg. Recall
99,6%	98,1%	95,1%	82,9%



Conclusions

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- We introduced Secure Routine paradigm to identify the vehicle's owner
- We provide FSparadigm for feature selection
- SR the best results in comparison with the other algorithms



Thanks for your attention

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